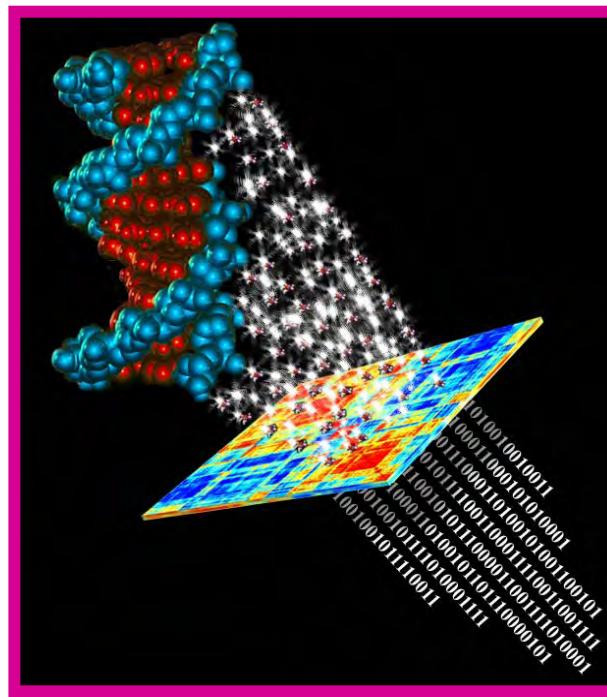


Integrative Pharm-Tox and the NCI-60: Genomics, Proteomics, and Bioinformatics



John N. Weinstein, M.D., Ph.D.
The Genomics & Bioinformatics Group
Laboratory of Molecular Pharmacology
Center for Cancer Research
National Cancer Institute



Center for
Cancer Research

Integrative Pharm-Tox and the NCI-60: Genomics, Proteomics, and Bioinformatics



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Study: Worldwide carbon dioxide emissions soar

Updated 56m ago | Comments 101 | Recommend 15

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By Dan Vergano, USA TODAY

Warnings about global warming may not be dire enough, according to a climate study that describes a runaway-train acceleration of industrial carbon dioxide emissions.

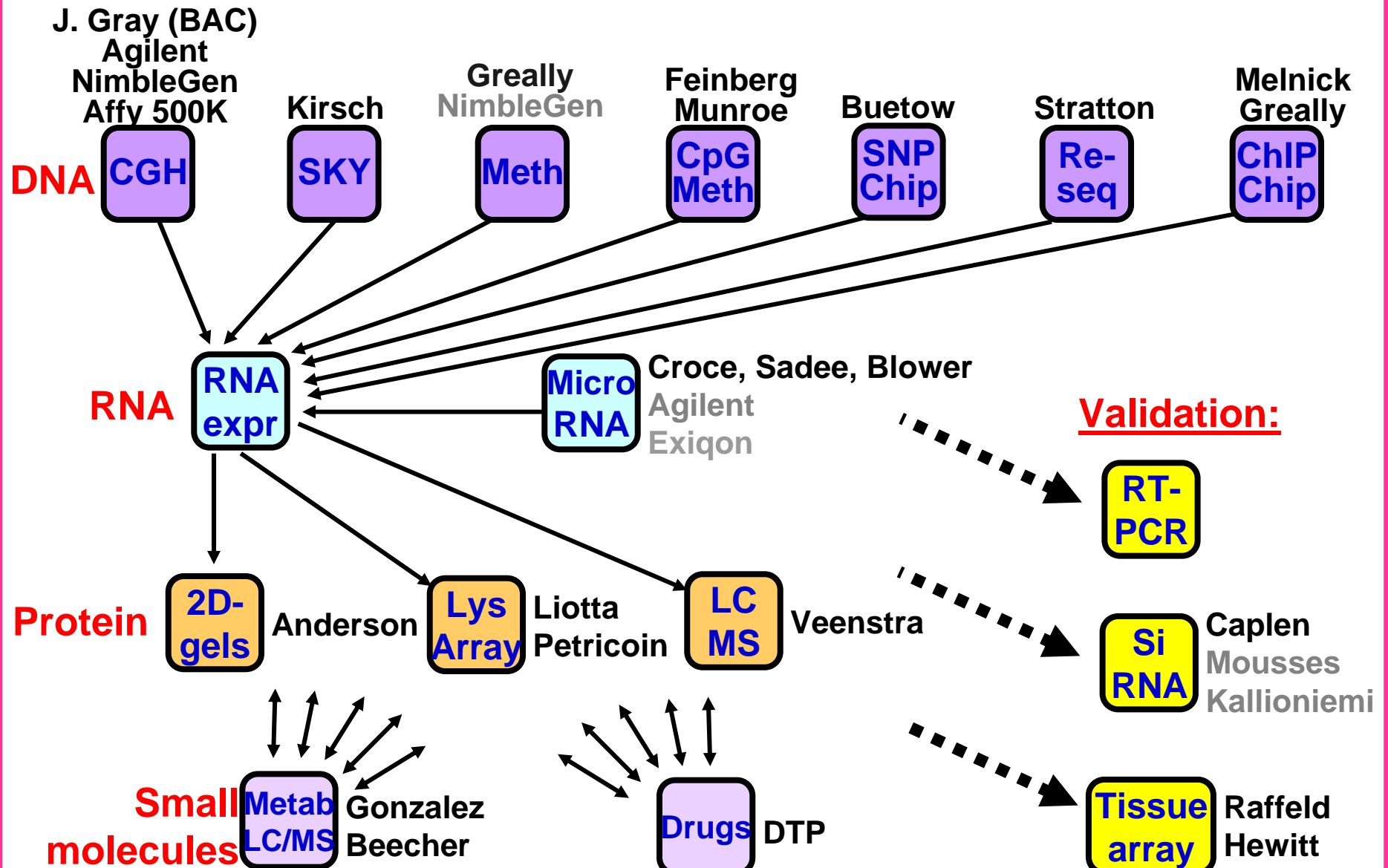
Fueled by rapid growth in coal-reliant China, rates of carbon dioxide emission from industrial sources increased from 2000 to 2004 "at a rate that is over three times the rate during the 1990s," says a report released by the journal *Proceedings of the National Academy of Sciences*.

Laboratory of Molecular Pharmacology
Center for Cancer Research
National Cancer Institute



Center for
Cancer Research

Integrative Molecular Profiling : The NCI-60, etc. (the GBG and Collaborators)



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Integrative Molecular Profiling : The NCI-60, etc. (the GBG and Collaborators)

J. Gray (BAC)

Agilent
NimbleGen
Affy 500K

Kirsch

Greally
NimbleGen

Feinberg
Munroe

Buetow

Stratton

Melnick
Greally

Integromics

(Weinstein & Pommier, Comptes Rendus Biol. 2003; 326:909)

ABC

Gottesman

The Cancer Genome Atlas (TCGA)

3D

GeneLogic

GeneLogic

Tax

The Cancer Genetic Markers of Susceptibility Project (CGEMS)

Center for
Cancer Research



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CellMiner: An Integrative Database/Tool for Molecular Profiling Data

Molecular Profile Data (transcript and protein profiles for TP53 in the NCI-60):

The screenshot shows the CellMiner interface with a blue header bar. The title "CellMiner™ Build:4" is at the top left. Below it is a navigation menu with links: Home, Cell Line Metadata, Dataset Metadata, Download Datasets, Query Datasets (highlighted in red), Data Intersection, Mutation Data, Credits, and Citing CellMiner. The main content area displays a table with data for TP53. The columns are labeled: Chip/Normalization Method, Probe Id, Gene Symbol, Chromosome, BR:MCF7, BR:MDA_MB_231, BR:HS578T, BR:BT_549, BR:T47D, CNS:SF_268, CNS:SF_295, and CNS. Two rows of data are shown: one for Protein: Lysate Array/Log2_condensed and another for RNA: cDNA.array/Log2.

Chip/Normalization Method	Probe Id	Gene Symbol	Chromosome	BR:MCF7	BR:MDA_MB_231	BR:HS578T	BR:BT_549	BR:T47D	CNS:SF_268	CNS:SF_295	CNS
Protein: Lysate Array/Log2_condensed	TP53_26_GBL00064	TP53	17	0.24	4.18	2.79	3.37	4.24	6.69	4.07	0.4
RNA: cDNA.array/Log2	2363338	TP53	17	0.803183718	0.926968296	0.366112977	1.178144685	0.316370002	1.262515144	0.018824549	0.125

At the bottom left are "Go to top" and "Back" buttons.

Cell Metadata (for the NCI-60 and drug-resistant cells):

The screenshot shows the CellMiner interface with a blue header bar. The title "CellMiner™ Build:4" is at the top left. To the right, it says "Genomics and Bioinformatics Group" and "LMP, CCR, National Cancer Institute". Below it is a navigation menu with links: Home, Cell Line Metadata (highlighted in red), Dataset Metadata, Download Datasets, Query Datasets, Data Intersection, Mutation Data, Credits, and Citing CellMiner. The main content area displays a table with cell metadata. The columns are labeled: Cell Name, Panel Name, Age, Sex, Prior Treatment, Tissue of Origin, Histology, Source, Ploidy, p53, mdr, and Doubling Time. Six rows of data are shown for different cell lines: CNS:SF_268, CNS:SF_295, CNS:SF_539, CNS:SNB_19, CNS:SNB_75, and CNS:U251.

Cell Name	Panel Name	Age	Sex	Prior Treatment	Tissue of Origin	Histology	Source	Ploidy	p53	mdr	Doubling Time
CNS:SF_268	CNS	24	F	Rad	non-epi	"Glioblastoma, ud"	NA	"2n+, Hyperdiploid (47-57)"	MT	-38	33.1
CNS:SF_295	CNS	67	F	Rad	non-epi	"Glioblastoma, ud"	NA	"5n+/-, Near-pentaploid 115+/- (104-126)"	MT	91	29.5
CNS:SF_539	CNS	34	F	Rad/BCNU/5FU/HU/6MP	non-epi	Glial cell neoplasm	NA	"4n+/-, Near-tetraploid 92+/- (81-103)"	WT	-40	35.4
CNS:SNB_19	CNS	47	M	None	non-epi	"Glioblastoma_ud"	NA	"3n+/-, Near-triploid 69+/- (58-80)"	MT	-41	34.6
CNS:SNB_75	CNS	NA	F	None							
CNS:U251	CNS	75	M	NA							

A blue box highlights the text "(Shankavaram, et al., Mol. Cancer Ther., 2007)" at the bottom right of the table.

Launched in the November, 2006 Issue



*American Association
for Cancer Research*

Molecular Cancer Therapeutics Introduces **Spotlight on Molecular Profiling**

A new series of pharmacogenomic and pharmacoproteomic articles focused on the NCI-60 cell line panel, highlighting the power and promise of “integromic” molecular profiling for therapeutics and for cancer patients in the post-genomic era.

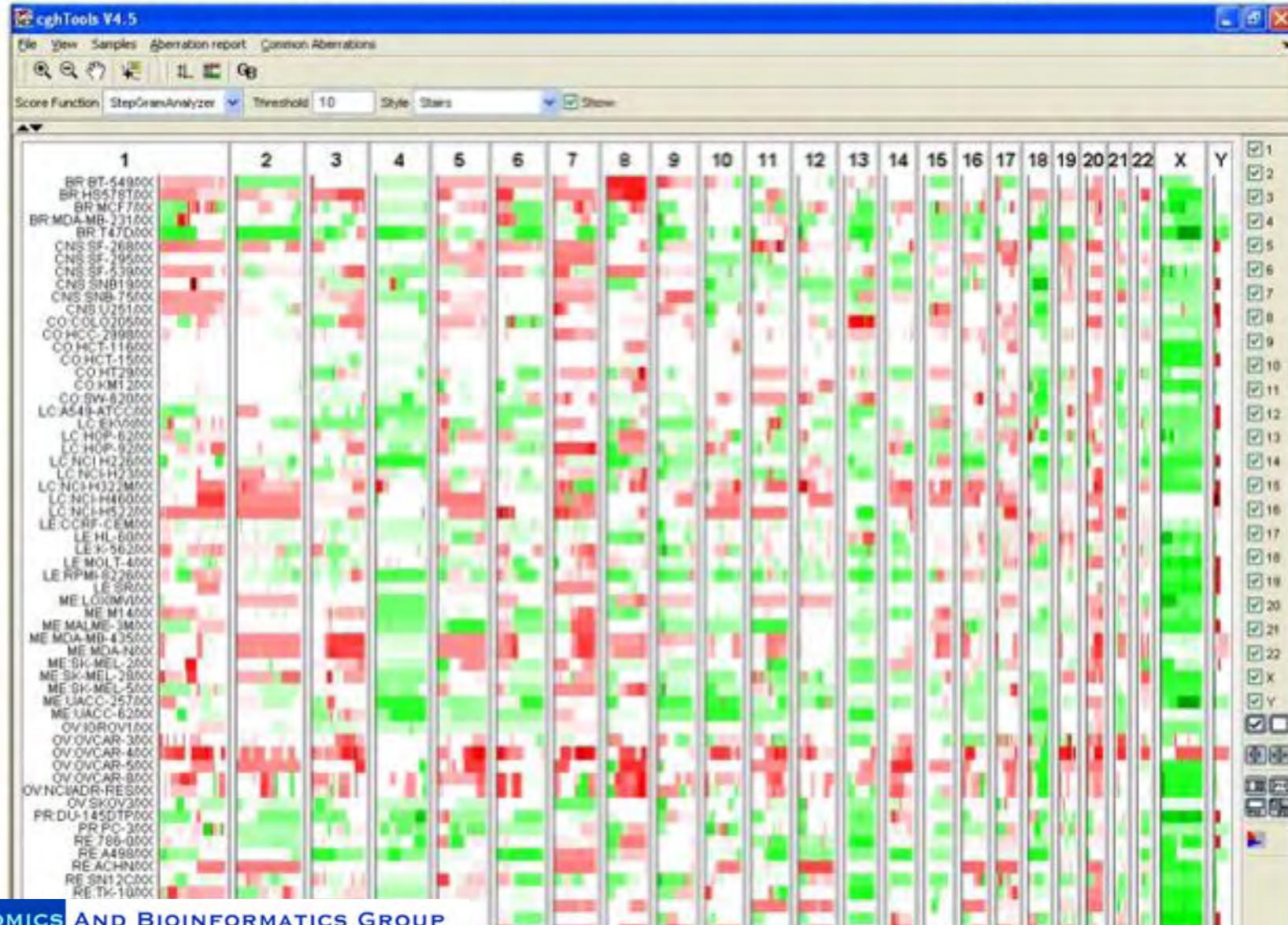
Now available FREE online at: <http://mct.aacrjournals.org>

High-Resolution Array CGH of the NCI-60

William Reinhold, Kimberly Bussey, Uma Shankavaram, John Weinstein

(Genomics & Bioinformatics Group, NCI)

Amir Ben-Don, Anya Tsalenko, Laurakay Bruhn, Zohar Yakhini, Doron Lipson, Michael Barrett (Agilent)

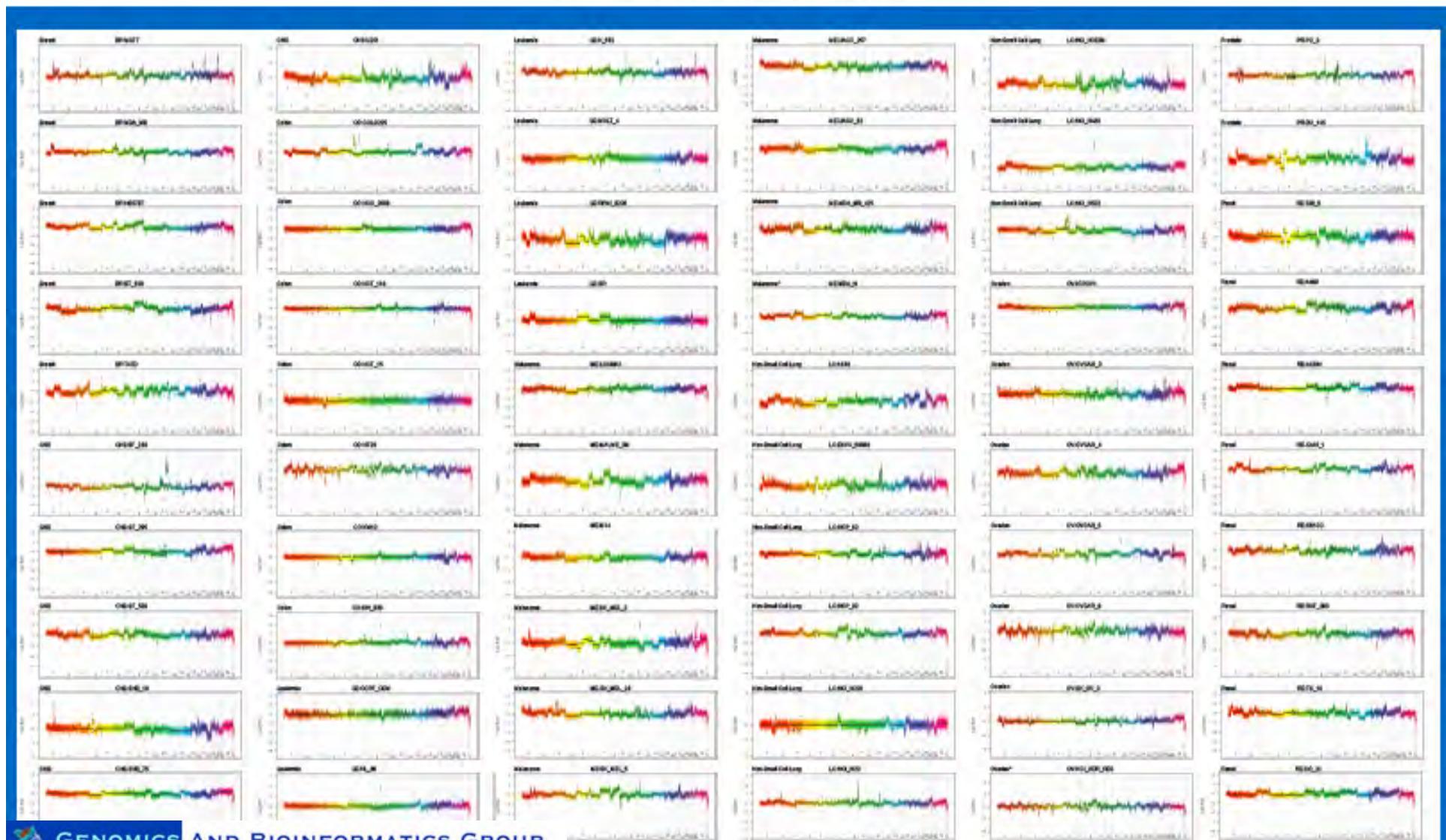


High-resolution oligonucleotide array CGH analysis of the NCI-60 cell

John N. Weinstein(1), William C. Reinhold(1), Kimberly J. Bussey(1), Mark Reimers(1), Uma Shankavaram(1),
Rebecca R. Selzer(2), Todd A. Richmond(2), Peggy S. Eise(2).

(1) Center for Cancer Research, National Cancer Institute, National Institutes of Health, Bethesda, MD.

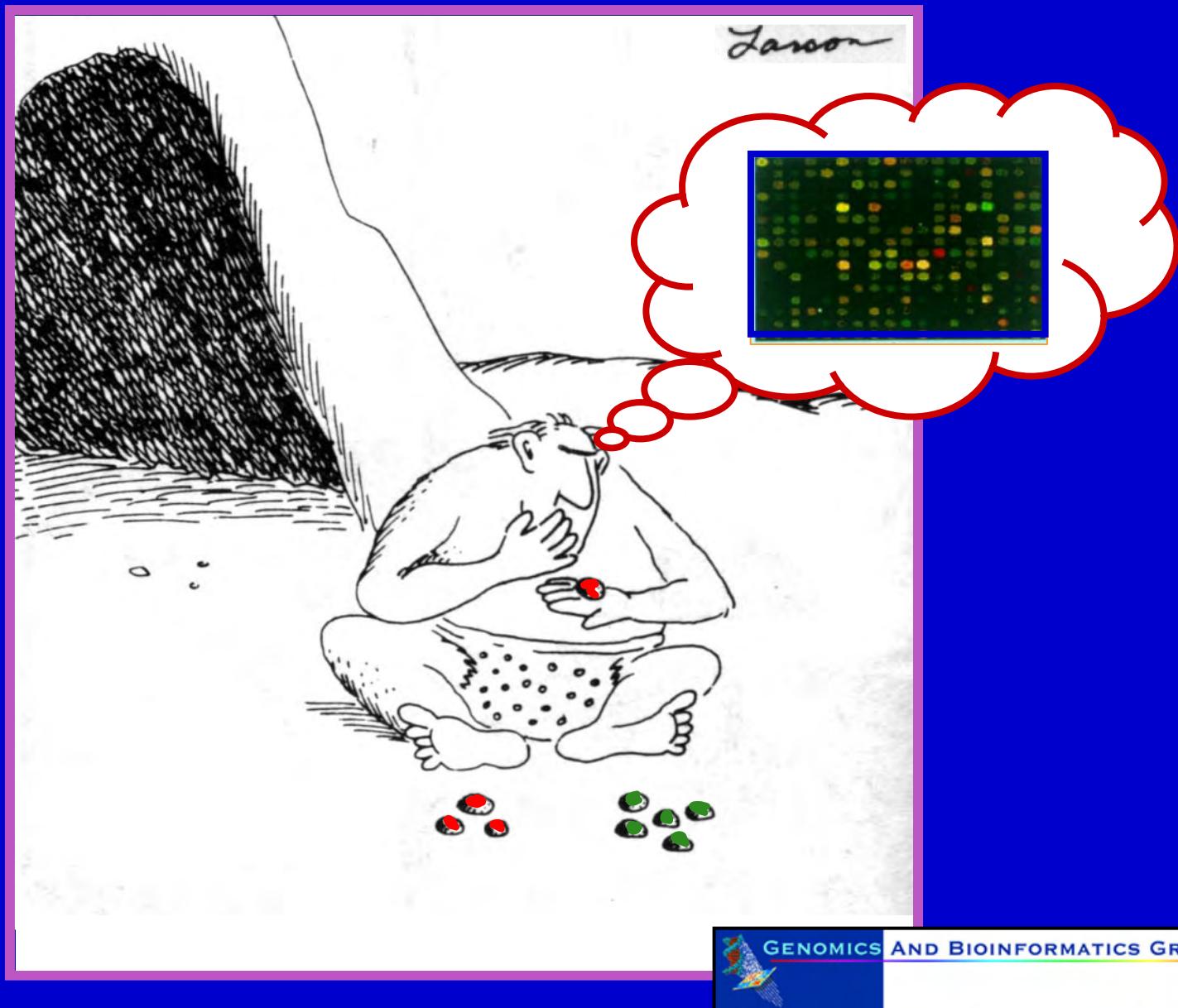
(2) NimbleGen Systems Inc., One Science Ct., Madison, WI.



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NimbleGen Systems, Inc.

Analyzing and Integrating Omic Data:

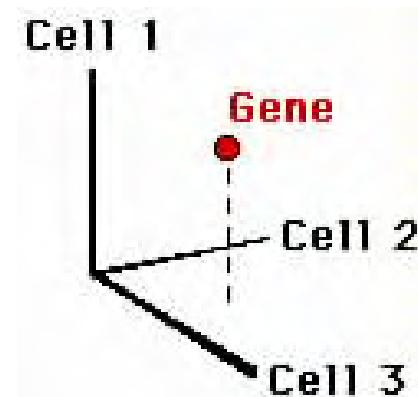


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Statistics of High-Dimensional Omic Data

From Classical Statistics

- Multivariate linear regression
- Principal component analysis
- Principal component regression (PCR)
- Multidimensional Scaling
- Partial least squares regression
- Stepwise regression
- ROC analysis
- SAM, PAM, Gene Shaving
- Correlation (Pearson, Spearman)
- Cluster analysis (e.g., hierarchical)
- Clustered image map (e.g., Weinstein, et al., **Science 275:343,1997**)

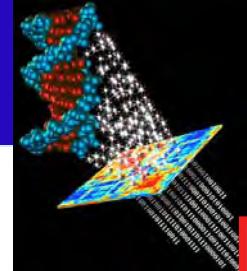


Computer-intensive statistical methods

- Bootstrap, Jackknife cross-validation
- Monte Carlo data randomizations, permutation tests
- Hidden Markov chain Monte Carlo
- Hierarchical error modeling with Gibbs sampling

From Machine Learning

- Neural networks (Weinstein, et al., **Science 258:447,1992**)
- Kohonen self-organizing maps (SOMs)
- Genetic function approximation
- Random Forest



Freely Available Bioinformatic Resources from the Genomics & Bioinformatics Group, LMP, CCR, NCI

(Head: John N. Weinstein, M.D., Ph.D.; jw4i@nih.gov)

(URL:discover.nci.nih.gov -- highlighted in Science (Netwatch)

MatchMiner: translates among gene identifier types for lists of hundreds or thousands of genes. Includes: GenBank accession numbers, IMAGE clone IDs, common gene names, HUGO names, gene symbols, UniGene clusters, FISH-mapped BAC clones, Affymetrix identifiers, and chromosome locations. (Bussey, et al., Genome Biology 2003; 4:R27)

GoMiner and High-throughput GoMiner (HTGM) leverage the Gene Ontology for large lists of genes, for example in microarray experiments. HTGM deals with multiple microarrays in a study, providing calculations and CIM graphics to integrate the data over the entire set of experiments (Zeeberg, et al., Genome Biology 2003; 4:R28 and BMC Bioinformatics 2005:6; 168).

CIMminer: generates color-coded Clustered Image Maps (CIMs) ("heat maps") to represent "high-dimensional" data sets such as gene expression profiles. We introduced CIMs in the early 1990's. Clustering of the axes brings like together with like to create patterns of color. (Weinstein, et al., Science 1997; 275:343-349).

MedMiner: searches and organizes the biomedical literature on genes, gene-gene relationships, and gene-drug relationships. MedMiner speeds up 5-fold the capture and organization of literature from PubMed searches. (Tanabe, et al., BioTechniques 1999; 27: 1210).

AbMiner: relational database of commercially available antibodies and quality-control information on them.

LeadScope/ links molecular markers and the drug discovery process. It links gene expression profiles for **LeadMinerTM:** the NCI-60 (or other screening cell panel) to a set of 27,000 chemical substructure descriptors of compounds tested against the cells. (Blower, et al., Nature's Pharmacogenomics J. 2002; 2:259)

SmudgeMiner: introduces new algorithms and visualizations for detecting regional artefacts on microarrays and assessing their influence. (Reimers and Weinstein, BMC Bioinformatics 2005:6: 166)

MIMminer: electronic Molecular Interaction Maps (MIMs) (Kohn, et al., Mol.Biol. Cell 2006;17: 1)

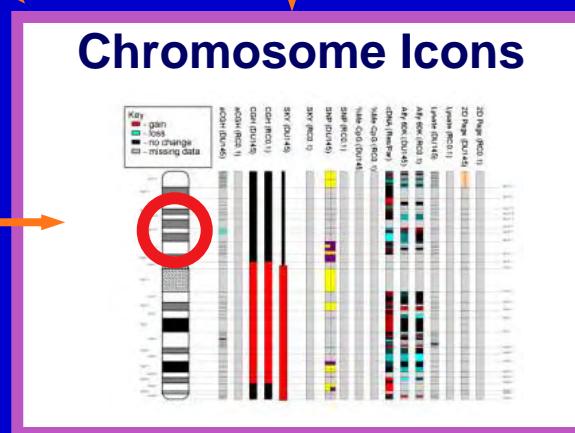
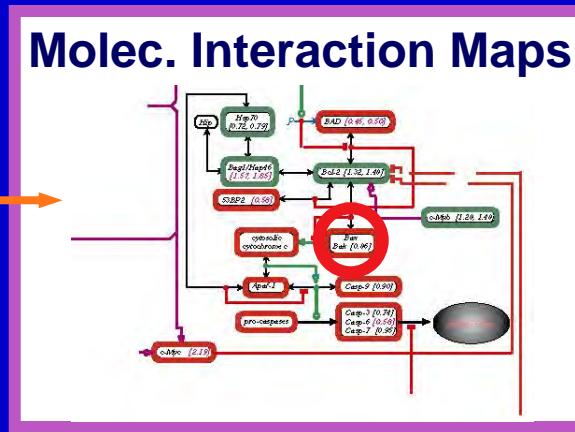
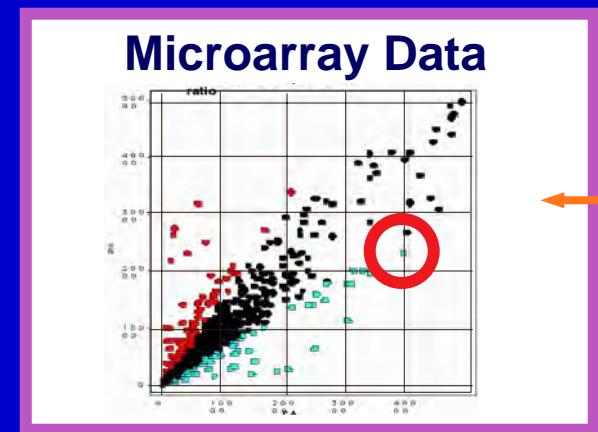
CellMiner: DNA, RNA, and protein molecular profiles on the NCI-60 and drug-resistant cells in a queryable, user-friendly web database/tool (Shankavaram, et al., Molec. Ca. Ther., in press).

SpliceMiner: Organizes information on splice variants of genes (Kahn, et al., BMC Bioinf. 2007;8:75).

<http://discover.nci.nih.gov> (lead software engineer: D. Kane, SRA)

The Miner Suite: Tools for Integromic Research

Input: DNA, RNA, Protein, Functional, Structural, Chemical, Pharmacological, Clinical Data



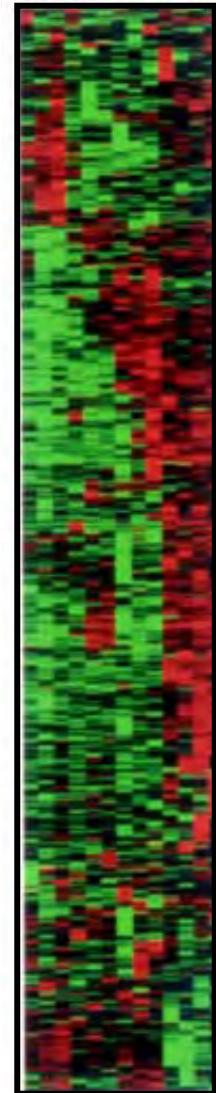
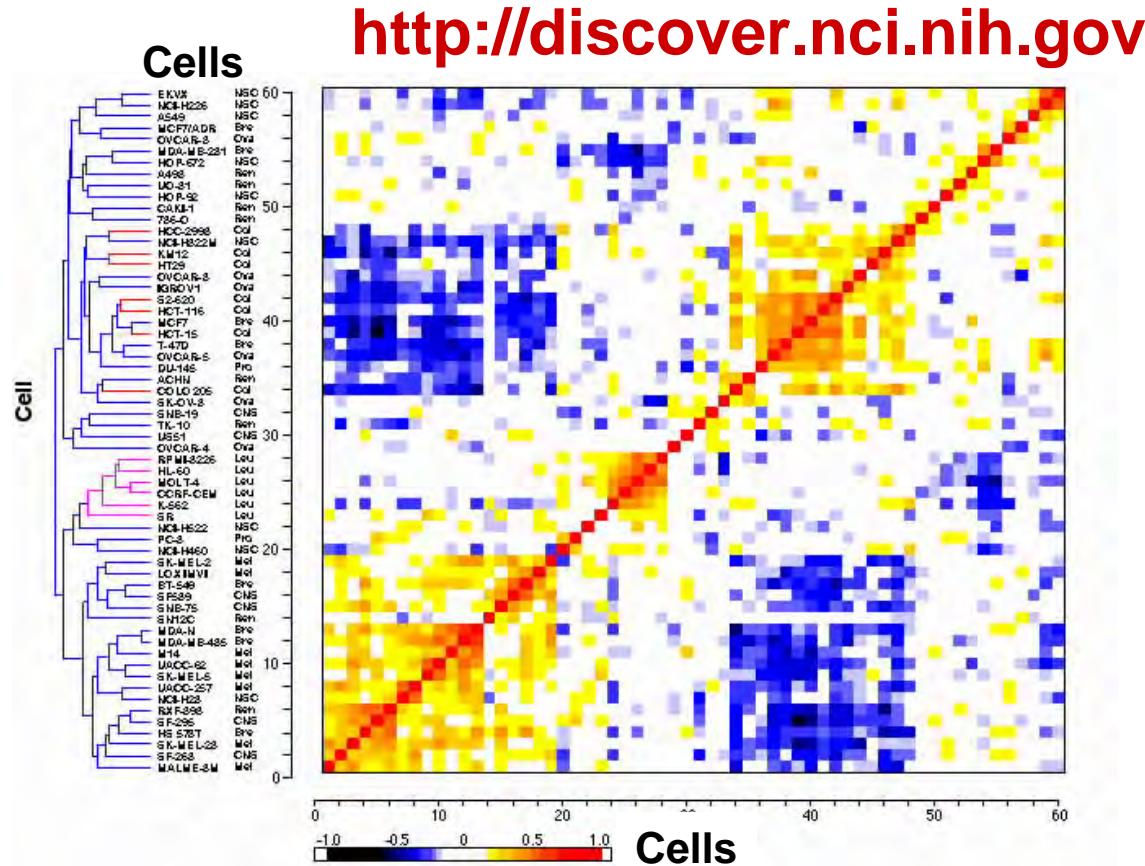
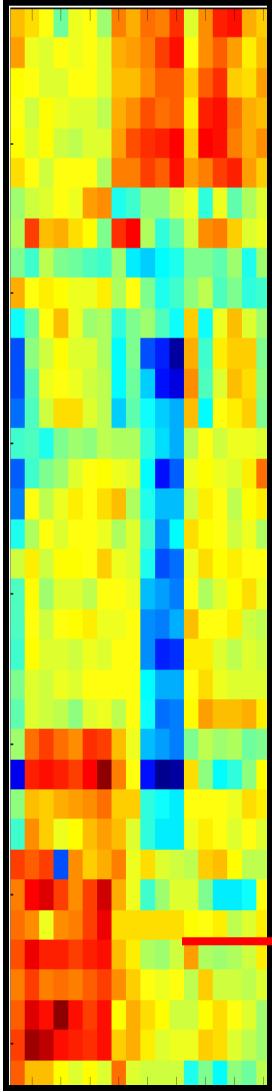
GBG Software:

- MIMminer
- CIMminer
- MatchMiner
- GoMiner
- MedMiner
- AbMiner
- LeadMiner
- MethMiner
- SmudgeMiner
- CellMiner
- SpliceMiner

External Resources:

- Spotfire
- GenBank
- GeneCards
- CGAP
- MAExplorer
- UniGene
- UCSC
- LocusLink
- OMIM

CIMminer: Clustered Image Maps (CIMs)



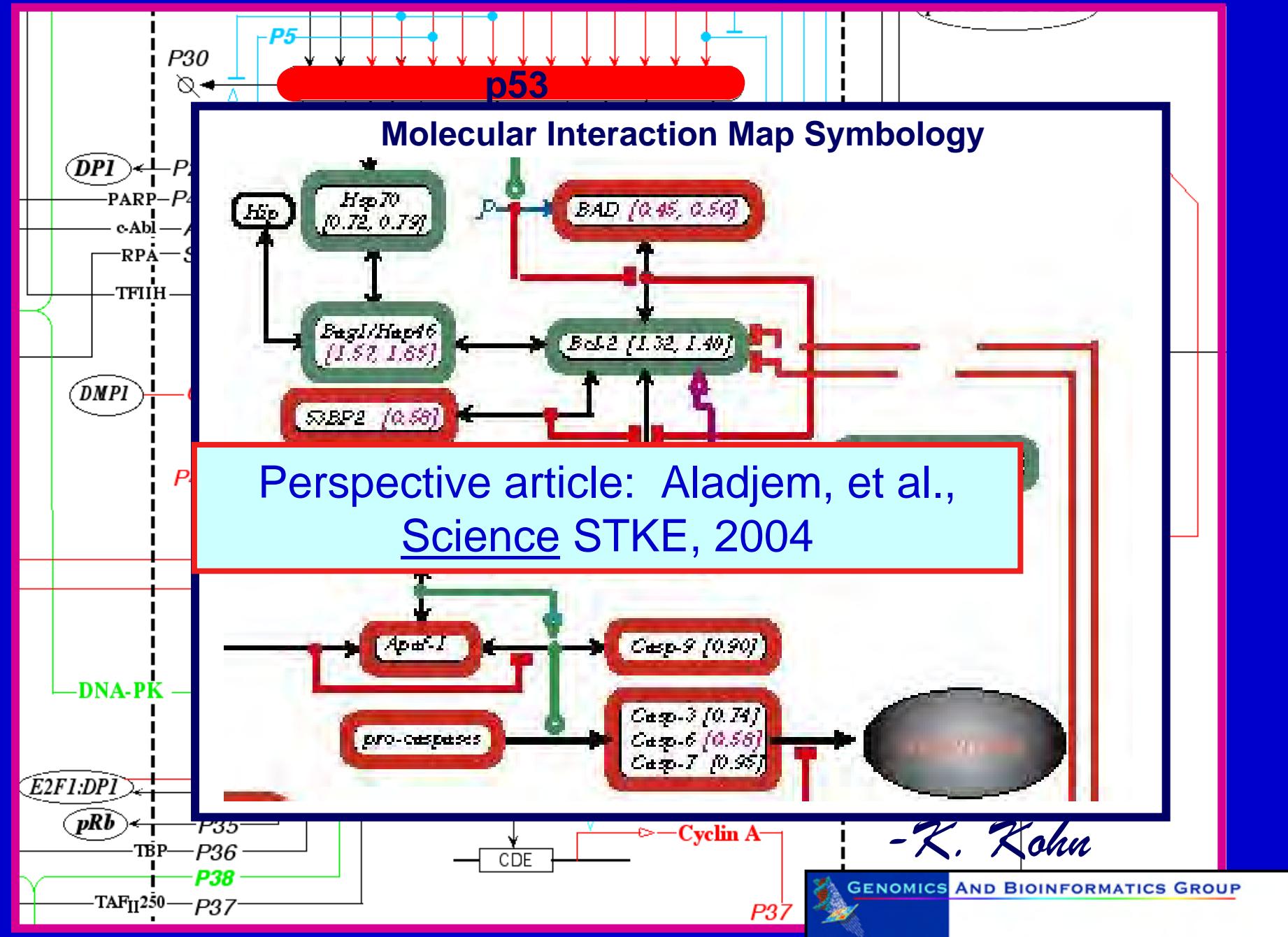
Weinstein et al, Stem Cells, 1994

Weinstein et al, Science, 1997

Myers...Weinstein, Electrophor., 1997

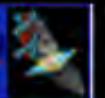
Eisen et al, PNAS, 1998

Electronic Molecular Interaction Map (eMIM) (<http://discover.nci.nih.gov>)



SpliceMiner Input

SpliceMiner


Genomics and Bioinformatics Group
LMP, CCR, National Cancer Institute 

[Home](#) [Query Database](#) [Microarray Tools](#) [FAQ](#) [Citing](#) [Credits](#)

[Interactive](#) [Batch](#)

Step 1: Select Organism

Step 2: Query Type

Gene Symbol
Example: BRCA1

OR

Genomic Position Chromosome:
Start Position:
Stop Position:
Example: Chromosome: 2 Start: 254956 Stop: 254976

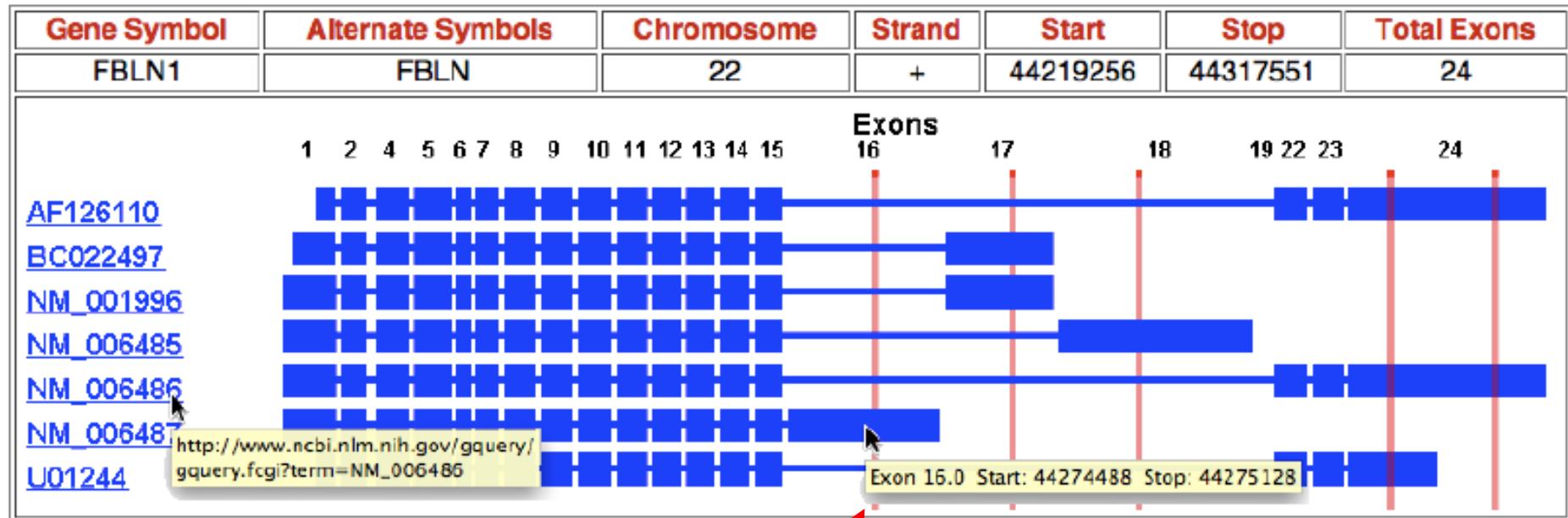
OR

Probe Sequence Sequence:
Example: GCGCAGAGGGGGGGGAGAGACACCCCCCTTC

Step 3: Submit your query

(Kahn, Zeeberg, et al., BMC Bioinf., 2007)

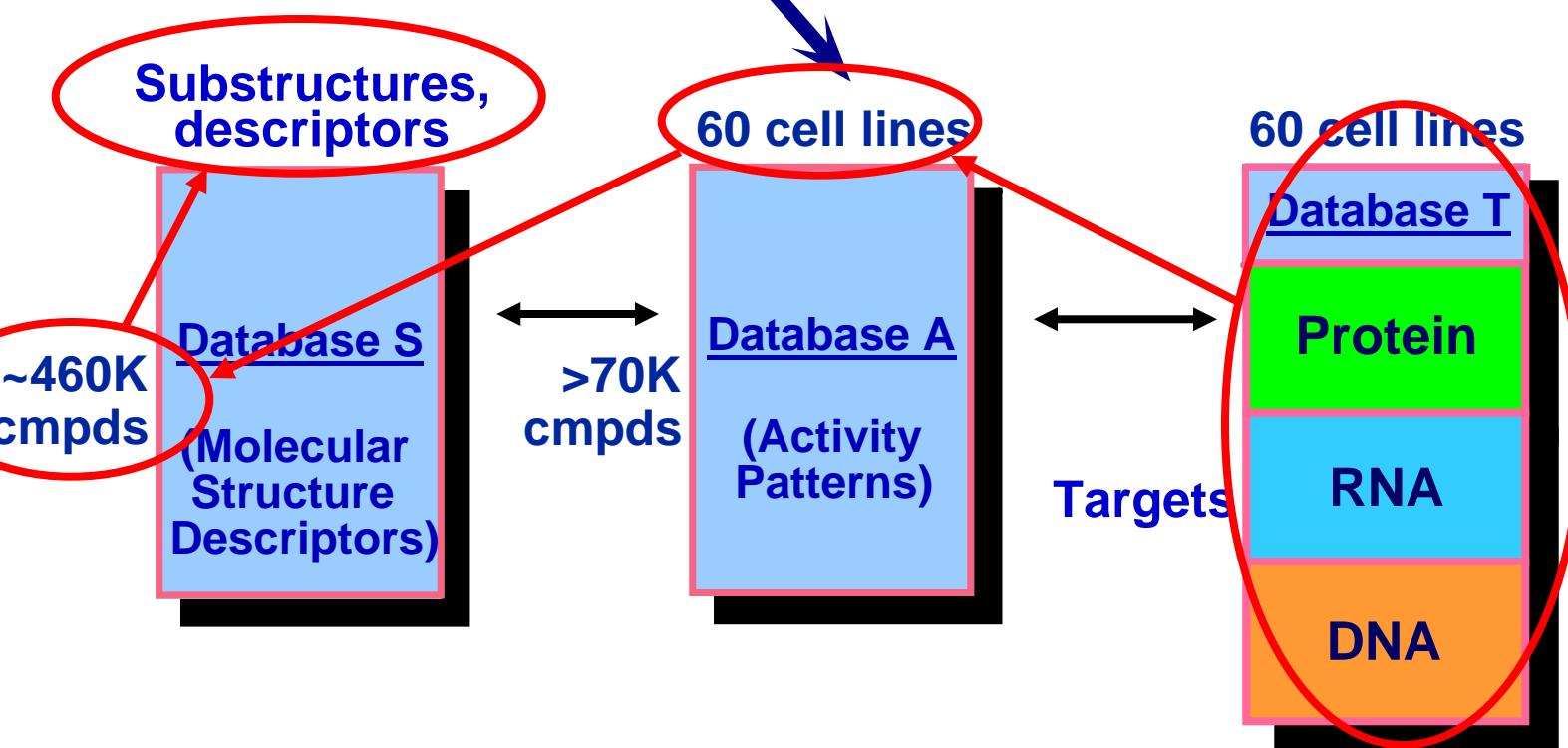
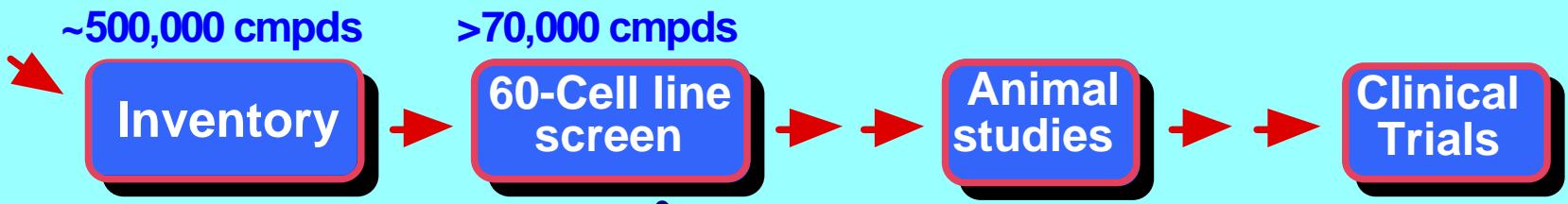
SpliceMiner Output



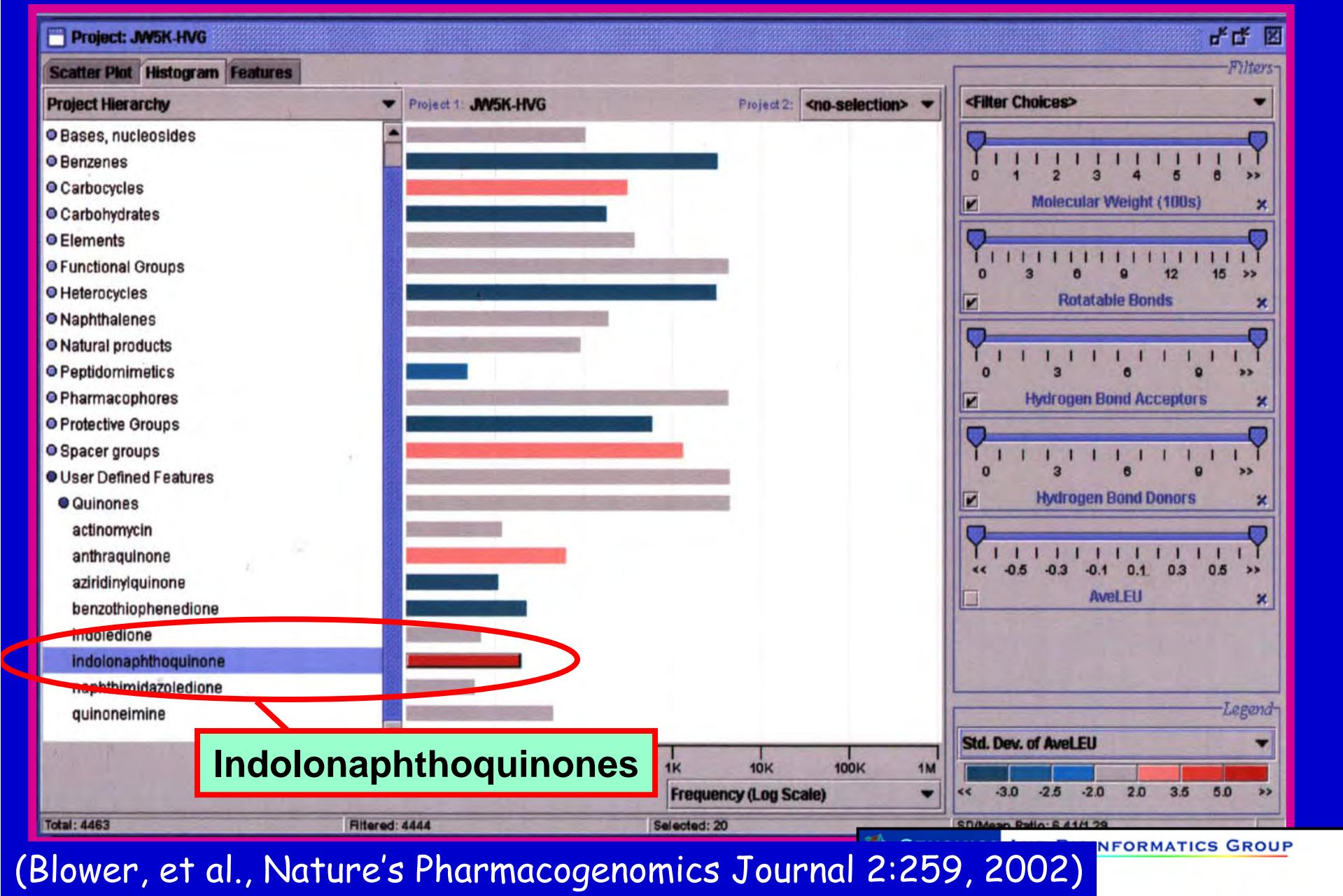
Probe

LeadsScope/LeadMiner: From Genes to Chemical Substructures

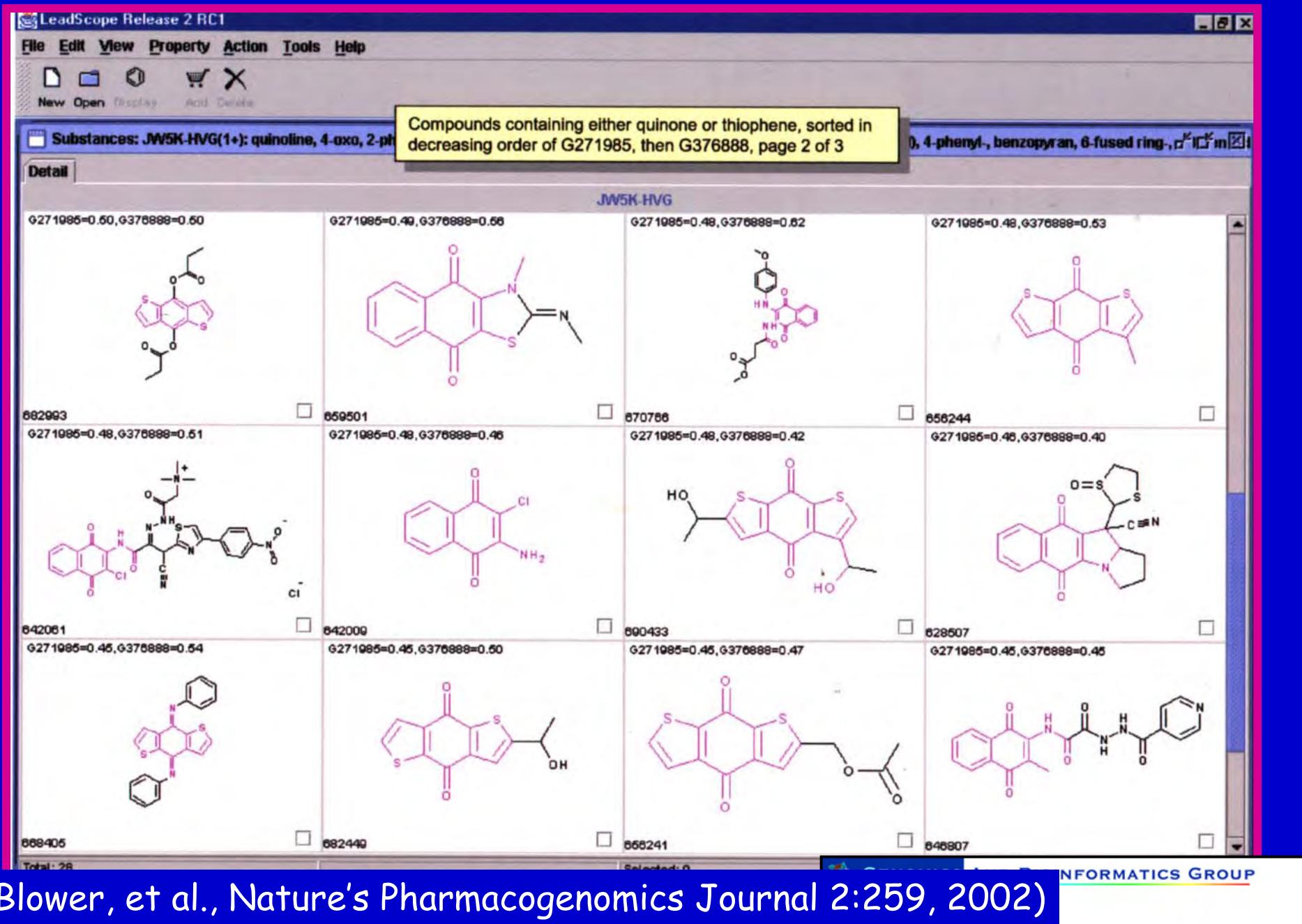
Flow in one portion of NCI Drug Discovery Program



LeadsScope/LeadMiner: From Genes to Chemical Substructures



LeadsScope/LeadMiner: From Genes to Chemical Substructures



Integromics: Translational Vignettes

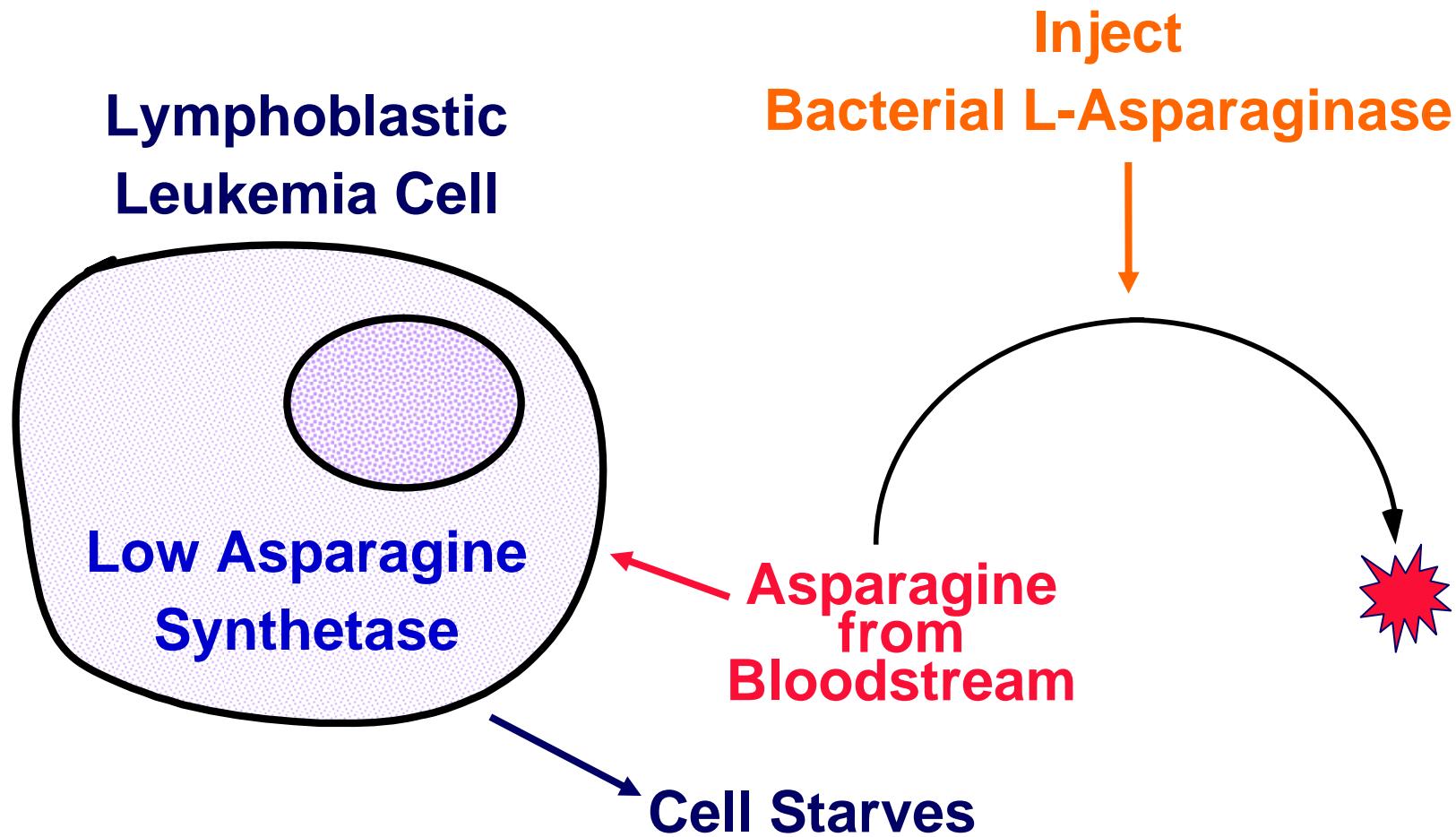


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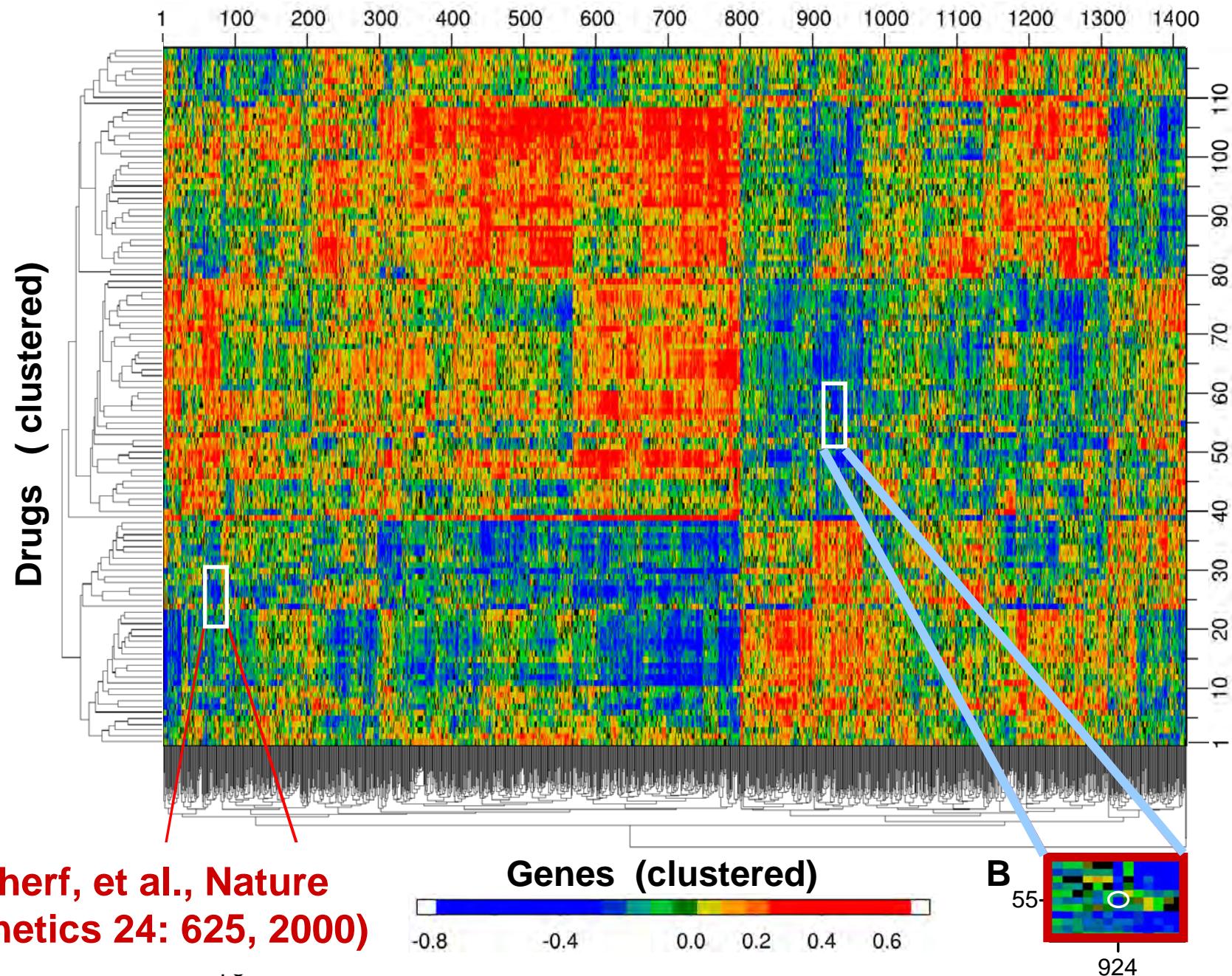
Integromic Vignette 1: Pharmacogenomics

Asparagine Synthetase Expression and L-Asparaginase Activity

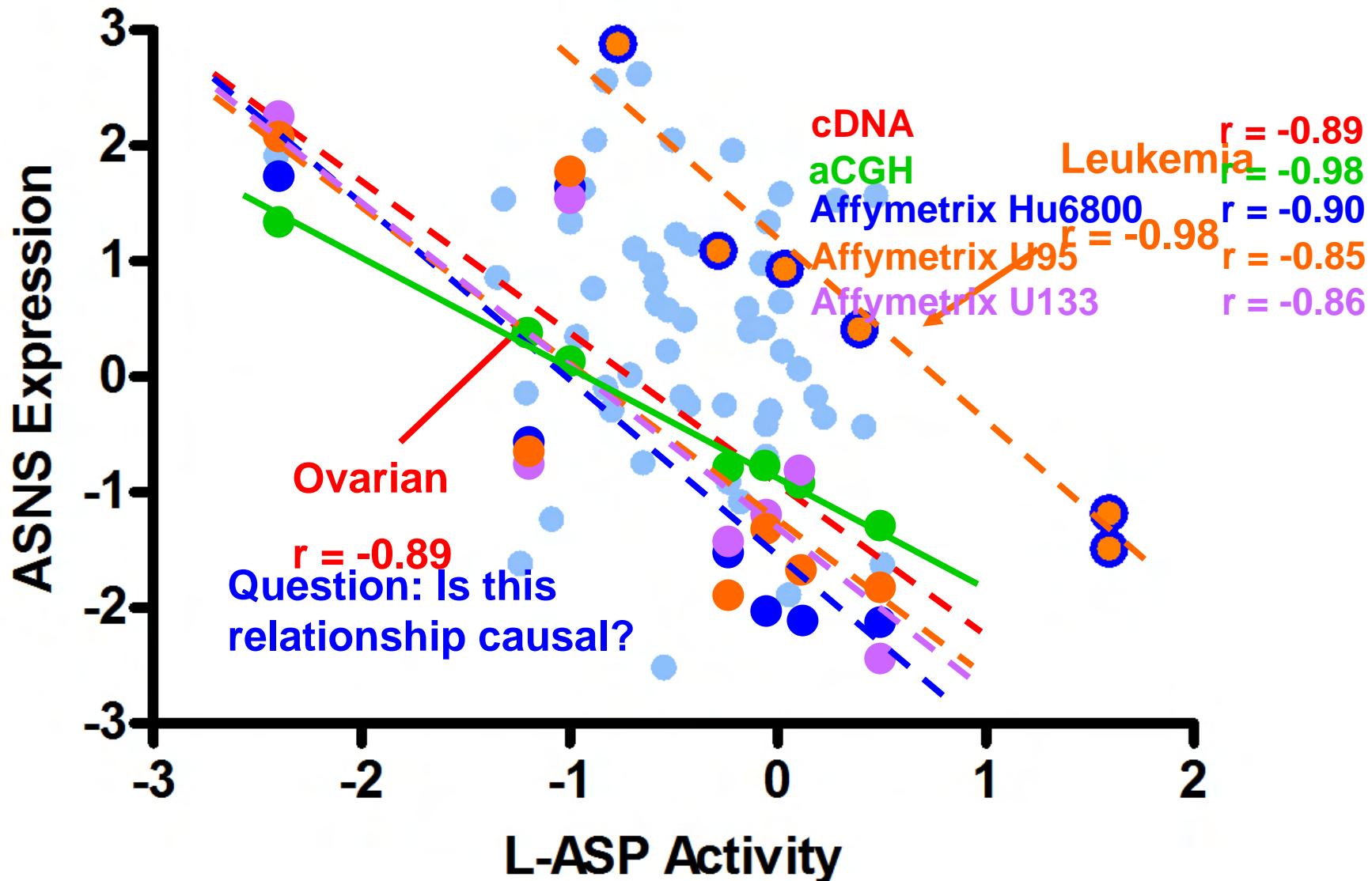
Rationale for Therapy with L-Asparaginase



Correlating Gene Expression Pattern with Drug Activity (CIM)

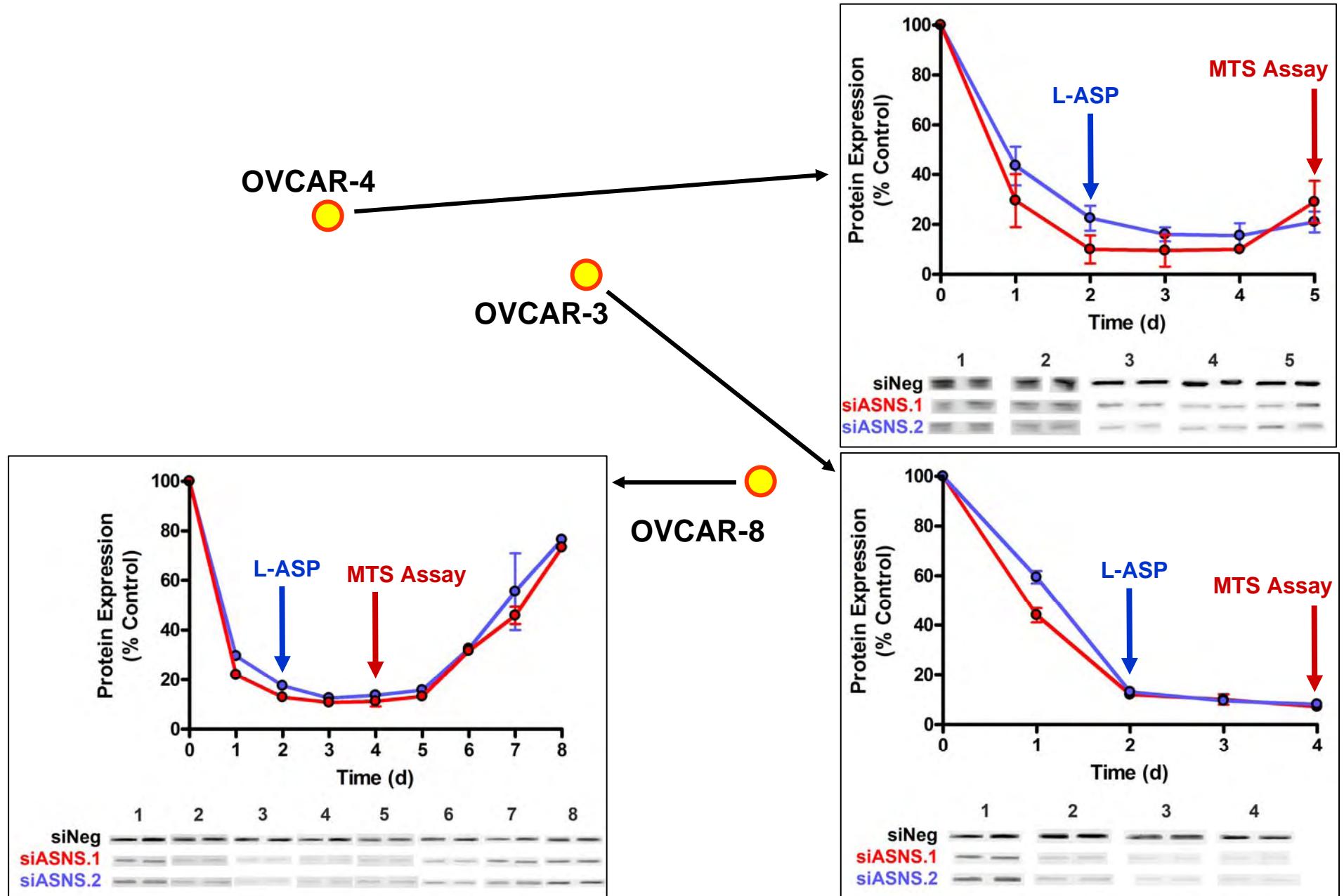


ASNS Expression vs. L-ASP Activity in the NCI-60 Cell Lines

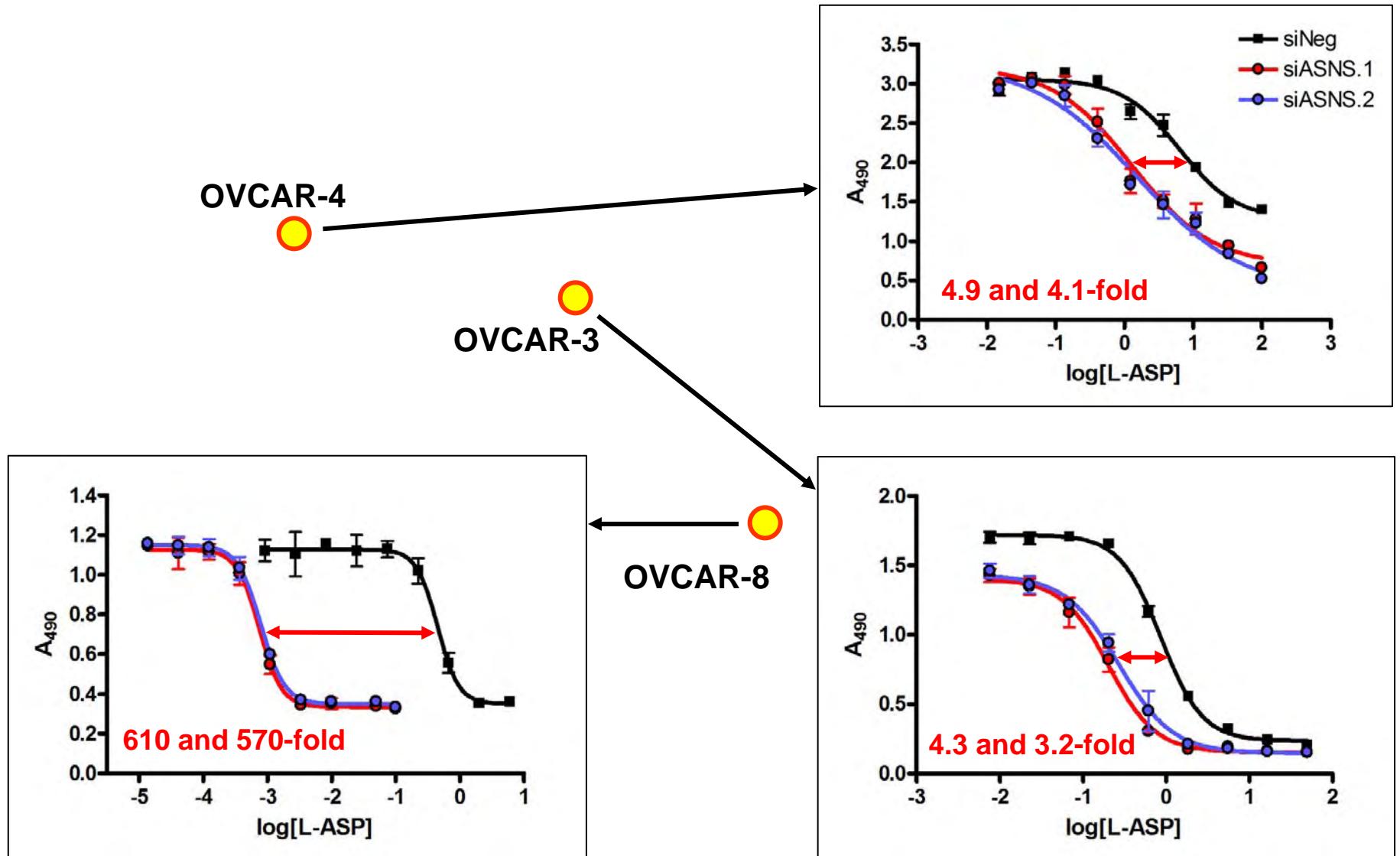


Scherf et al., *Nat Genet* (2000); Bussey et al., *Mol Ca. Ther* (2006); Lorenzi et al., *Mol Cancer Ther*, submitted

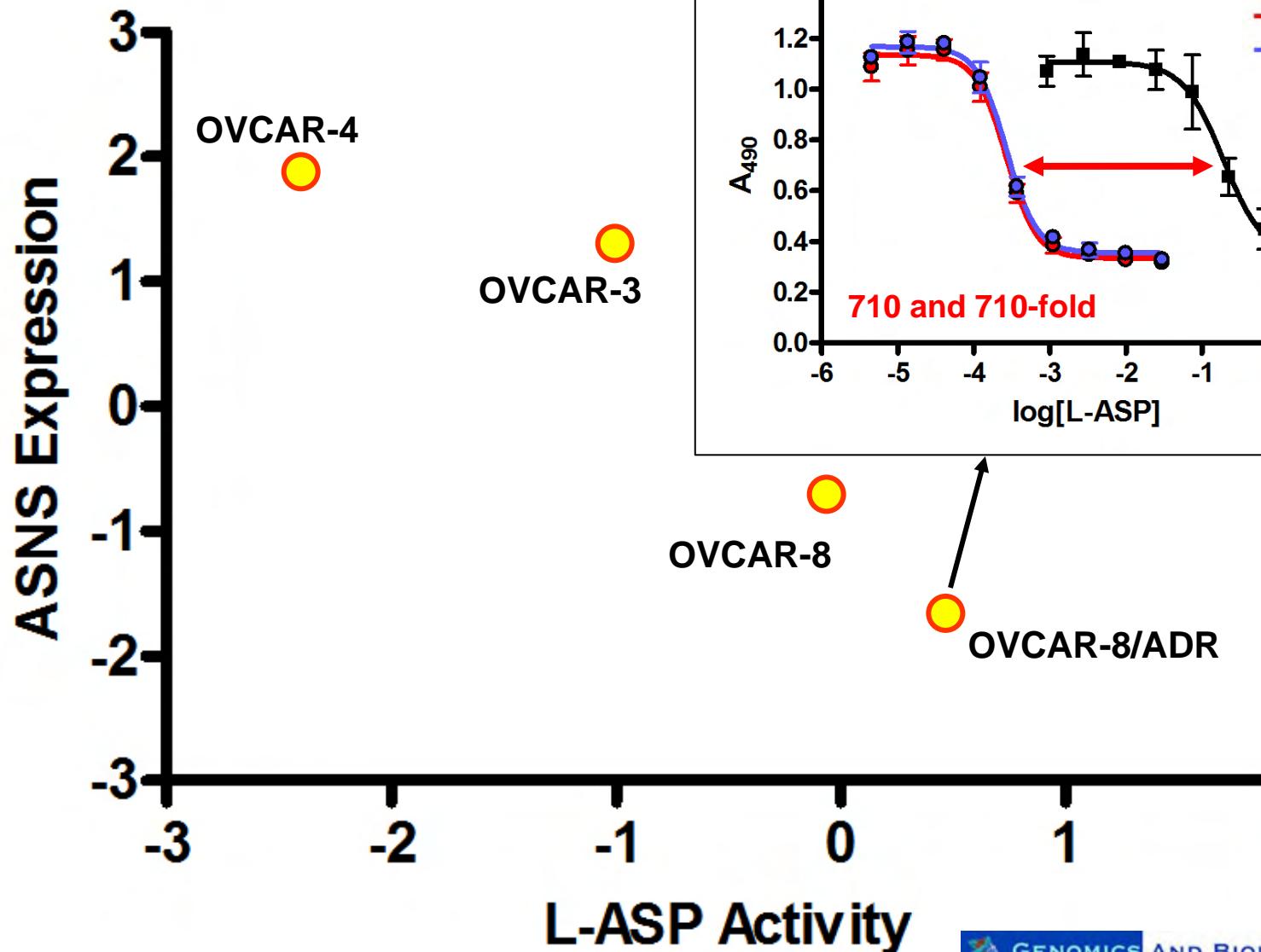
ASNS Protein Expression after siRNA Transfection



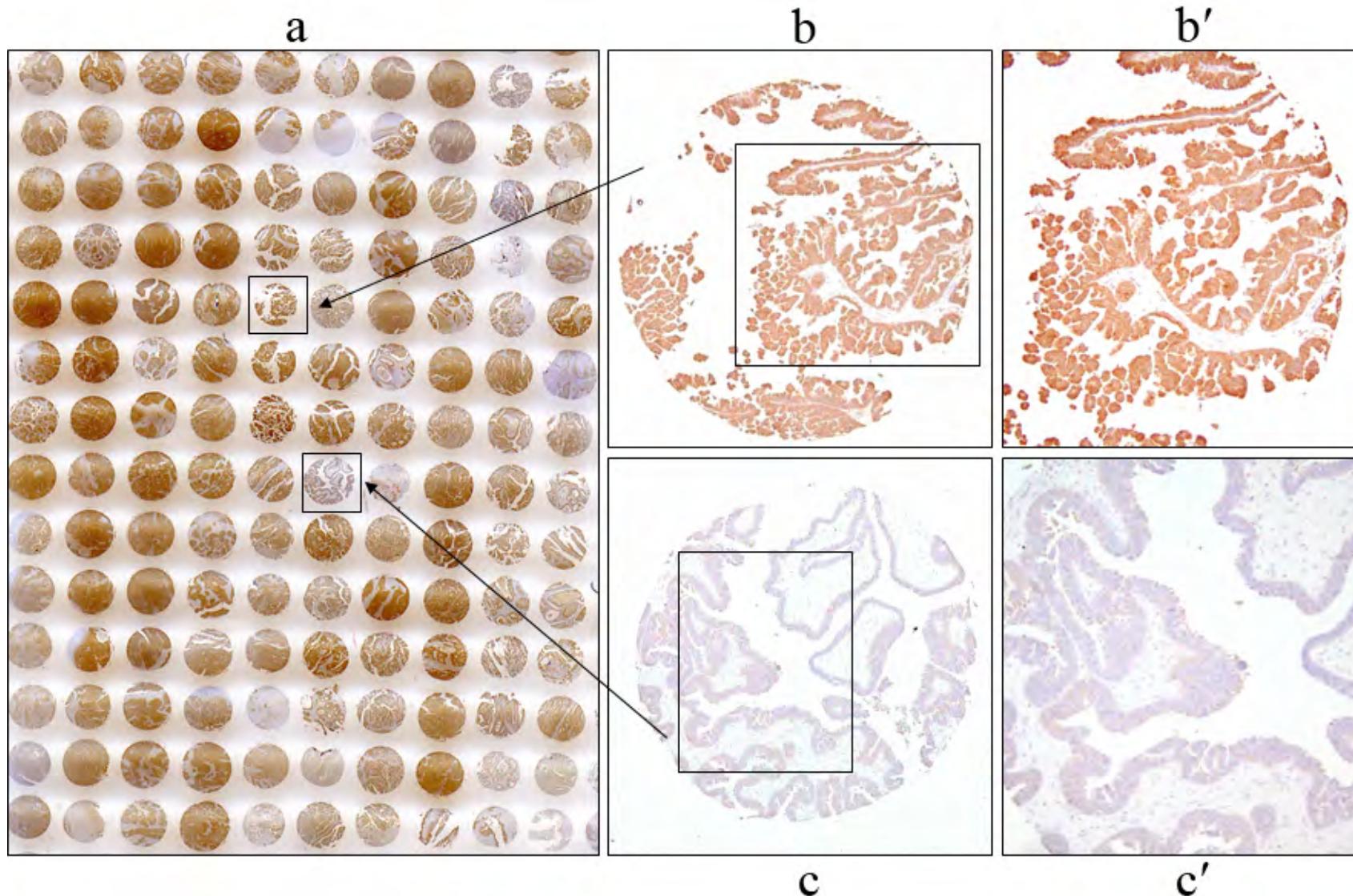
Response to L-ASP After siRNA Knock-down



Multidrug Resistance and the ASNS/L-ASP Relationship



Ovarian Cancer Tissue Array



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Conclusions: L-Asparaginase

- There is a pharmacogenomic / pharmacoproteomic relationship between L-ASP activity and the expression of ASNS in ovarian cancer cell lines.
 - That relationship is causal (siRNA expts).
 - The relationship is not affected by classical multidrug resistance.
 - The findings provide rationale for clinical testing of L-ASP against low-ASNS subsets of ovarian cancers with ASNS as a causally linked biomarker.
-
-

Omics - Beyond Genomics and Proteomics

- Genomics
- Proteomics
- Transcriptomics
- Kinoetics
- CHOomics
- Immunomics
- Toxicogenomics
- Metabolomics
- Metagenomics
- Translational genomics
- Clinical genomics

Synergy between omic

and hypothesis-driven research

Synergy between experimental science and bioinformatics

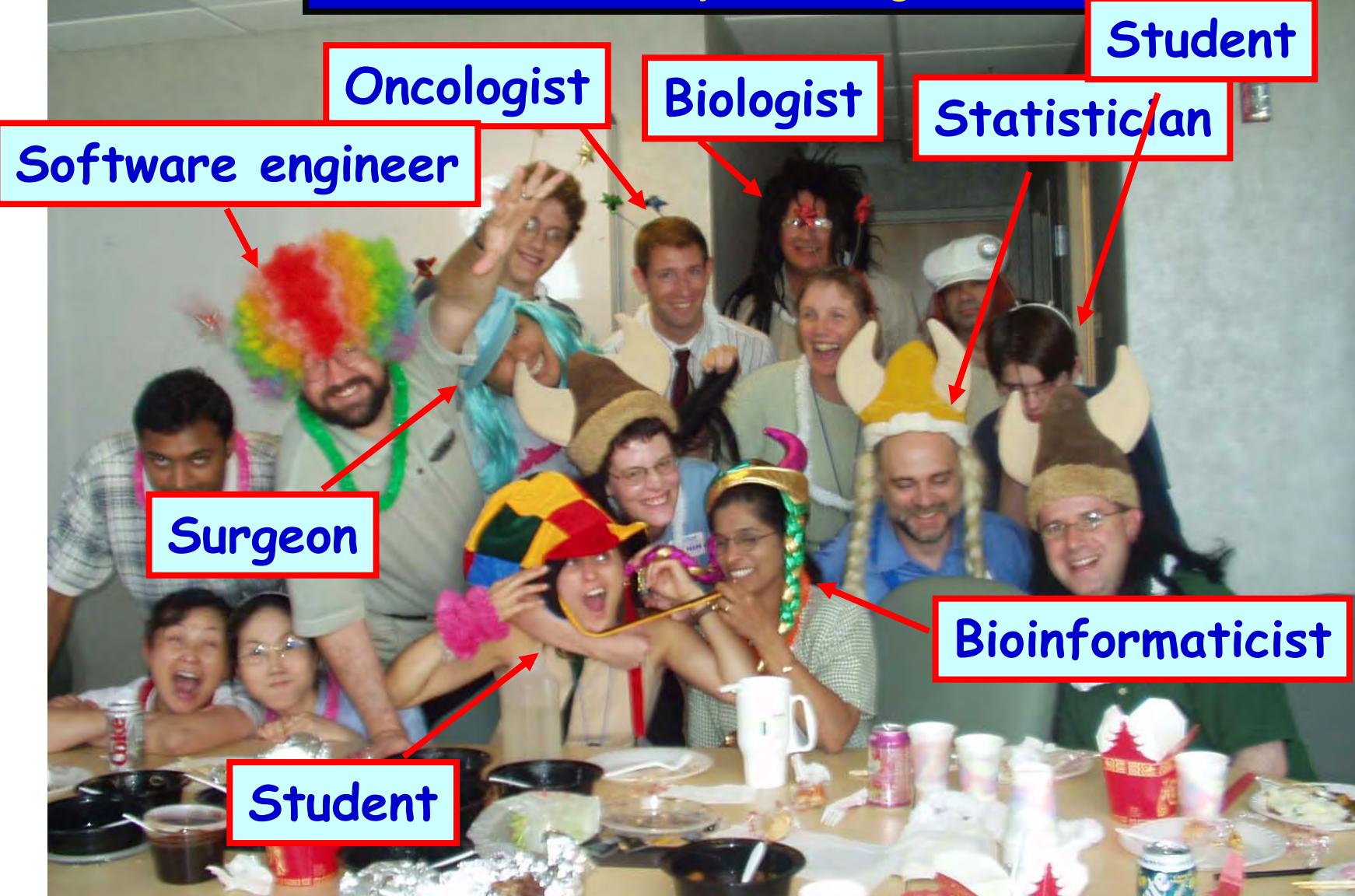
Synergy among scientific cultures

Synergy between domain and analytical expertise

Synergy among types of molecular data

Integromics™

Genomics & Bioinformatics Group Laboratory Meeting



Principal NIH Collaborators

Genomics & Bioinformatics Group

- John Weinstein
- William Reinhold
- Gabriel Eichler
- Philip Lorenzi
- Ogechi Ikediobi
- Barry Zeeberg

Past Members

- Uma Shankavaram
- Satoshi Nishizuka
- Mark Waltham
- Samir Lababidi
- Kim Bussey
- Jes Alexander
- Hosein Kouros-Mehr
- Alika Maunakea
- Jae Lee
- Fuad Gwadry
- Sohyoung Kim
- Sylvia major
- Uwe Scherf
- Ari Kahn
- Sudar Narasimhan (GBG, SRA)
- Ike Nnabuggwu (GBG, SRA)
- Ajay

GBG Computer Science Team

- David Kane (GBG, SRA)
- Margot Sunshine (GBG, SRA)
- Sohana Choudhry(GBG, SRA)
- Hong Cao (GBG, SRA)

LMP, NCI

- Yves Pommier
- Kurt Kohn
- Mirit Aladjem

DTP, NCI

- Ken Paull
- Dominic Scudiero
- Anne Monks
- David Covell
- Susan Holbeck
- Dan Zaharevitz
- Mark Kunkel
- Michael Grever
- Michael Boyd
- Bruce Chabner

NIH

- N Caplen
- P Goldsmith
- E Kohn
- M Raffeld
- M Gottesman
- S Ambs
- C Harris
- T Fojo
- S Bates
- S Hewitt
- G Szakacs
- J-P Annereau
- R Klausner
- JC Barrett
- J Riss
- I Kirsch
- A Roschke
- K Buetow
- D Nelson
- L Rubinstein
- T Veenstra
- F Gonzalez
- G Alvord
- J Roayaeaei
- P Munson

Principal Collaborators Outside of NIH

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- D Ross, C. Perou, M Eisen,
P Brown, D Botstein

UCSF

- J Gray, K Chen

Broad

- E Lander, T Golub, J Staunton, J
Mesirov

Ohio State

- P Blower, W Sadee, Y Huang, C
Croce, CG Liu

NCI, GMU

- L Liotta, E Petricoin, C Jamison

Albert Einstein

- J Greally, A Melnick

Georgetown U.

- A Fornace, Hongfang Liu

VTT Center, Finland

- O Kallioniemi

TGen

- D von Hoff, S Mousses, M
Barrett

Large Scale Biology

- N L. Anderson

GeneLogic

- E Kaldjian, J. Cossman,
D Dolginow

Agilent

- Z Yakhini, D Lipson, A Ben-
Dor,

NimbleGen

- R Seltzer, P Eis, R Green

ExonHit

- R Einstein

Enzon

- I Horak

Metabolic Analyses

- C. Beecher